

(1)

# Orthogonal Polynomials

$w(x) > 0$  on  $I$

$\pi_i(x)$   $i=0,1,\dots$  degree  $i$

$P_i(x)$  = mono. orth. poly.

$$\langle \pi_i, \pi_j \rangle = \int_I \pi_i(x) \pi_j(x) w(x) dx = \delta_{ij}$$

$$\pi_i(x) = k_i x^i + \dots$$

$$P_i(x) = \pi_i(x) / k_i$$

$$\langle P_i, P_i \rangle = 1/k_i^2 \equiv h_i$$

$$A = \begin{bmatrix} \pi_0(x) & \pi_1(x) & \dots & \pi_{n-1}(x) \end{bmatrix}$$

$$D = \begin{bmatrix} w(x) \end{bmatrix}$$

Cauchy Binet:

$$\det(A^T D A) = \sum_{i_1 < i_2 < \dots} d_{i_1} d_{i_2} \dots A \begin{pmatrix} i_1 & i_2 & i_3 & \dots \\ 1 & 2 & 3 & \dots \end{pmatrix}^2$$

$$\rightarrow \int \pi w(x_i) \begin{vmatrix} \pi_{i_1}(x_i) & \dots & \pi_{i_n}(x_i) \\ \pi_{i_1}(x_n) & \dots & \pi_{i_n}(x_n) \end{vmatrix}^2 dx_1 \dots dx_n$$

$$= \frac{1}{n!} \int \pi w(x_i) \begin{vmatrix} | & | & | \\ | & | & | \\ | & | & | \end{vmatrix}^2 dx_1 \dots dx_n$$

$$\det \begin{bmatrix} \int_I \pi_i(x) \pi_j(x) w(x) dx \\ \vdots \\ \int_I \pi_{n-1}(x) \pi_j(x) w(x) dx \end{bmatrix} = \frac{1}{n!} \frac{h_{i_1} h_{i_2} \dots h_{i_n}}{k_{i_1} k_{i_2} \dots k_{i_n}} \int \pi w(x_i) \prod (x_i - x_j)^2 dx_1 \dots dx_n$$

$0 \leq i, j \leq n-1$

(2)

$$\int_{x_1, \dots, x_n} \prod v(x_i) \prod |x_i - x_j|^2 dx_1 \dots dx_n = n! h_1 h_2 \dots h_{n-1}$$

Let  $w(x) = \sqrt{\frac{\alpha}{\pi}} e^{-\alpha x^2}$  (normal with  $\sigma^2 = \frac{1}{2\alpha}$ )

$\frac{1}{n! h_1 \dots h_{n-1}} \left(\frac{\alpha}{\pi}\right)^{n/2} e^{-\alpha(\sum x_i^2)} \prod |x_i - x_j|^2$  is a probability density on  $\mathbb{R}^n$

$$\int_{h_1 \dots h_{n-1}} 1 = \frac{2^{n(n-1)/2}}{\prod_{i=1}^{n-1} \Gamma(i)}$$

$$Pr(\text{all } x_i \leq y) = \frac{2^{n(n-1)/2}}{n! \prod_{i=1}^{n-1} \Gamma(i)} \int_{-\infty, y}^{\infty} \int_{-\infty, y}^{\infty} \left(\frac{\alpha}{\pi}\right)^{n/2} e^{-\alpha \sum x_i^2} \prod |x_i - x_j|^2 dx_1 \dots dx_n$$

$$\int_{h_1(\alpha, \infty) \dots h_{n-1}(\alpha, \infty)} \dots$$

$$Pr(\text{all } x_i \leq y) = \frac{\prod_{i=0}^{n-1} h_i(\alpha, y)}{\prod_{i=0}^{n-1} h_i(\alpha, \infty)} \leftarrow w(x) = e^{-\alpha x^2} \text{ on } (-\infty, \infty)$$
$$\leftarrow w(x) = e^{-\alpha x^2} \text{ on } \mathbb{R}$$

(3)

Nothing really depends on  $w(x)$

One can always define  $w_{ij}(x) = \begin{cases} w(x) & x \leq y \\ 0 & x > y \end{cases}$

Obtain  $\pi_i(x; y)$  with  $h_i(x; y)$

$$p(x_1, \dots, x_n) = \frac{1}{n! \prod_{k=0}^{n-1} h_k(x_k)} \prod_{k=0}^{n-1} \pi_k(x_k) \quad \text{is a probability on } \mathbb{R}^n$$

$$Pr(\text{all } x_i \leq y) = \frac{\prod_{k=0}^{n-1} h_k(x_i; y)}{\prod_{k=0}^{n-1} h_k(x_i; \infty)}$$

Given  $w(x)$  how do you compute  $h_i$ ?  
Generally?

I believe best is to get  $T$  as fast as possible:

$$T = \begin{pmatrix} \alpha_1 & \beta_1 & & & \\ & \beta_1 & & & \\ & & \beta_2 & & \\ & & & \beta_{n-1} & \\ & & & & \alpha_n \end{pmatrix}$$

$$x \pi_k = \alpha_{k+1} \pi_k + \beta_k \pi_{k-1} + \beta_{k+1} \pi_{k+1}$$

If  $\int_I w(x) dx = 1$  then  $\pi_0(x) = 1$

$$T \begin{pmatrix} \pi_0(x) \\ \pi_1(x) \\ \vdots \\ \pi_{n-1}(x) \end{pmatrix} = x \begin{pmatrix} \pi_0(x) \\ \pi_1(x) \\ \vdots \\ \pi_{n-1}(x) \end{pmatrix}$$

$$\begin{aligned} & \beta_{k-1} \pi_{k-2}(x) \\ & + \alpha_k \pi_{k-1}(x) \\ & + \beta_k \pi_k(x) \\ & = x \pi_{k-1}(x) \end{aligned}$$

U

$$k_i = \beta_1 \dots \beta_{i+1} = \sqrt{h_i}$$

$$h_i = \frac{1}{\beta_1^2 \dots \beta_{i+1}^2}$$

How to compute T:

Moment Method (ill conditioned)

$$m_k = \int_I x^k x$$

$$H = \begin{bmatrix} m_0 & m_1 & m_2 \\ m_1 & m_2 & \vdots \\ m_2 & \vdots & \vdots \\ \vdots & \vdots & \vdots \end{bmatrix}$$

Hankel

Cholesky  $R^T R = H$

$R =$  

$$\beta_k = \frac{R_{k+1,k+1}}{R_{k,k}}$$

$$\alpha_k = \frac{R_{k+1,k+2}}{R_{k+1,k+1}} - \frac{R_{k,k+1}}{R_{k,k}}$$

(5)

$$\begin{aligned}x \pi_k &= \alpha_{k+1} \pi_k + \beta_k \pi_{k-1} + \beta_{k+1} \pi_{k+1} \\x \pi_i &= \alpha_{i+1} \pi_i + \beta_i \pi_{i-1} + \beta_{i+1} \pi_{i+1}\end{aligned}$$

$$\pi_i = k_i r_i$$

$$x k_i r_i = \alpha_{i+1} k_i r_i + \beta_i k_{i-1} r_{i-1} + \beta_{i+1} k_{i+1} r_{i+1}$$

$$\begin{aligned}P_{i+1}(x) + \frac{\alpha_{i+1} k_i}{\beta_{i+1} k_{i+1}} P_i(x) + \frac{\beta_i k_{i-1}}{\beta_{i+1} k_{i+1}} P_{i-1}(x) \\= \frac{x k_i}{\beta_{i+1} k_{i+1}} P_i(x)\end{aligned}$$

$$\frac{k_i}{\beta_{i+1} k_{i+1}} = 1$$

$$\cancel{S_i} = \cancel{D_{i+1}}$$

$$P_{i+1} + \underbrace{\alpha_{i+1}}_{S_i} P_i + \underbrace{\frac{\beta_i k_{i-1}}{k_i}}_{R_i} P_{i-1} = x P_i$$

$$P_{i+1} + \alpha_{i+1} P_i + \beta_i \beta_{i+1} P_{i-1} = x P_i$$

(6)

$$R_L = \frac{h_i}{h_{n-1}} = \frac{1}{\beta_{n+1}^2}$$

$$\langle \pi_{n+1}, \pi_n \rangle = \delta_{n+1}$$

$$\int_{-\infty}^y x P_n^2(x) e^{-x^2} dx = \frac{1}{2} \frac{d}{dy} \int_{-\infty}^y P_n^2(x) e^{-x^2} dx = \int_{-\infty}^y P_n^2(x) e^{-x^2} dx$$

$$\int_{-\infty}^y P_n^2(x) e^{-x^2} dx$$

$$\frac{d}{dy} \left[ \int_{-\infty}^y P_n^2(x) e^{-x^2} dx \right] = P_n^2(y) e^{-y^2} + \int_{-\infty}^y 2 P_n(x) \frac{\partial P_n(x)}{\partial y} e^{-x^2} dx$$

$$dW P_n^2 e^{-x^2}$$

$$u = x$$

7

$$d\alpha = y e^{-dy^2}$$

$$v = -\frac{e^{-dy^2}}{2d}$$

$$u = P_n^2(y)$$

$$du = 2P_n P_n'$$

$$\int u dv = \int P_n^2(y) \cdot 2P_n P_n' dy$$

$$= \frac{P_n^3(y)}{3} + \int \cancel{2P_n P_n'} dy$$